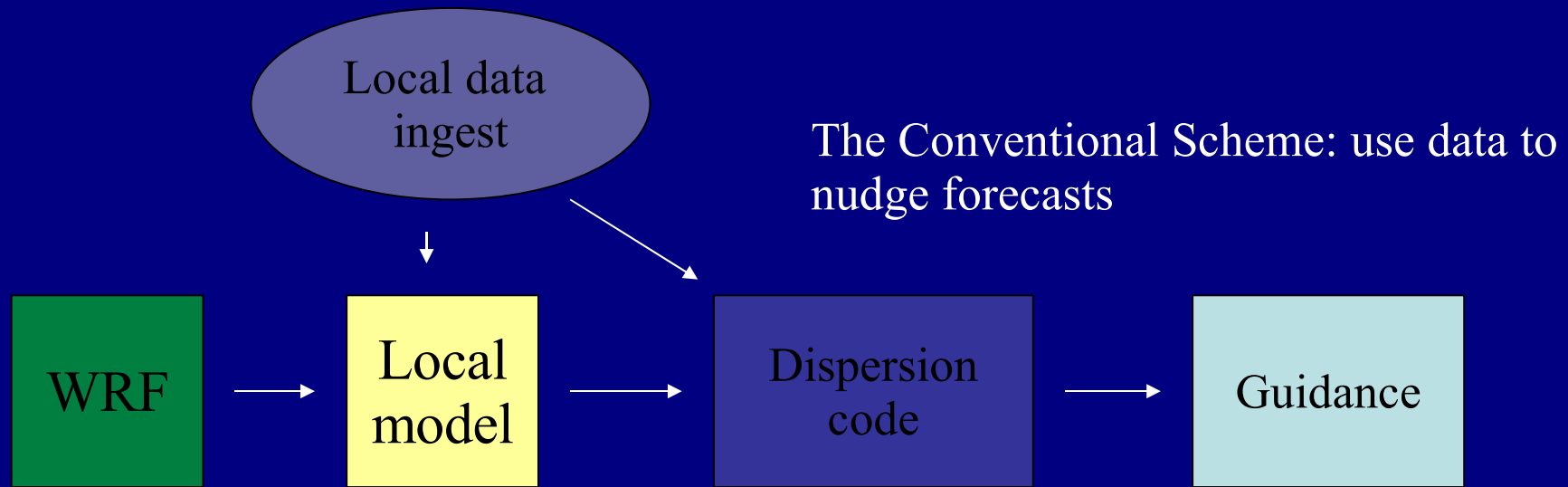


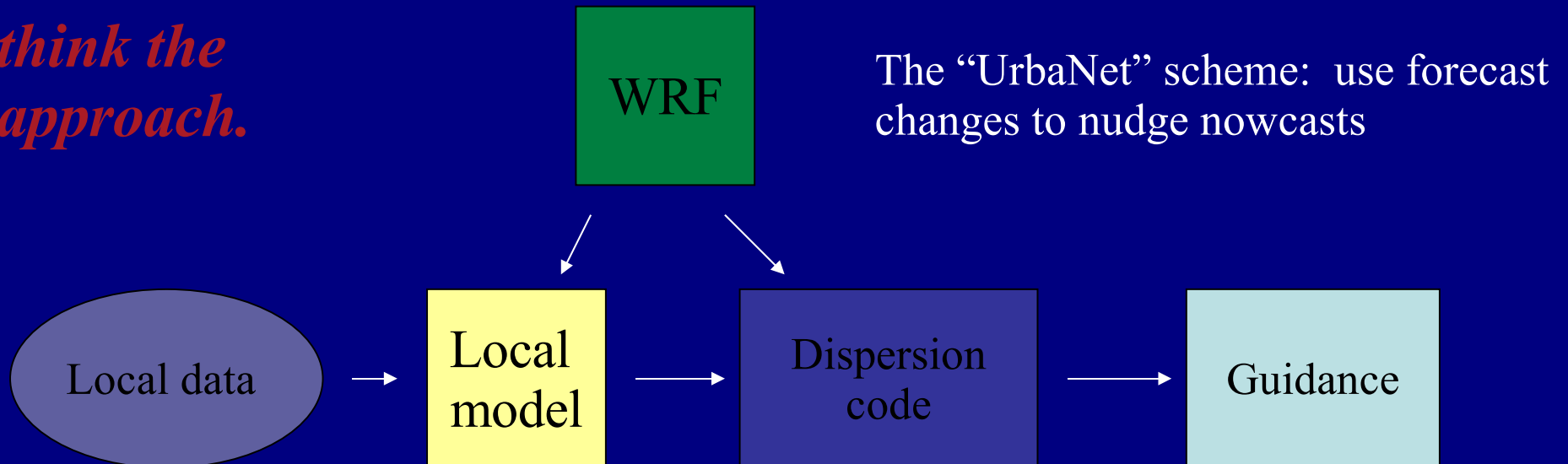
Urban Model Output Statistics

Richard M. Eckman
NOAA/ARL Field Research Division
Idaho Falls



This works well on the average. For events, it often fails.

We need to re-think the approach.



Probabilistic Forecast Uncertainty

F = Quantity we want to predict

M = Model estimate of quantity

O_a = Observations used to initialize model

$$p(F|O_a) = \int p(F|M) p(M|O_a) dM$$

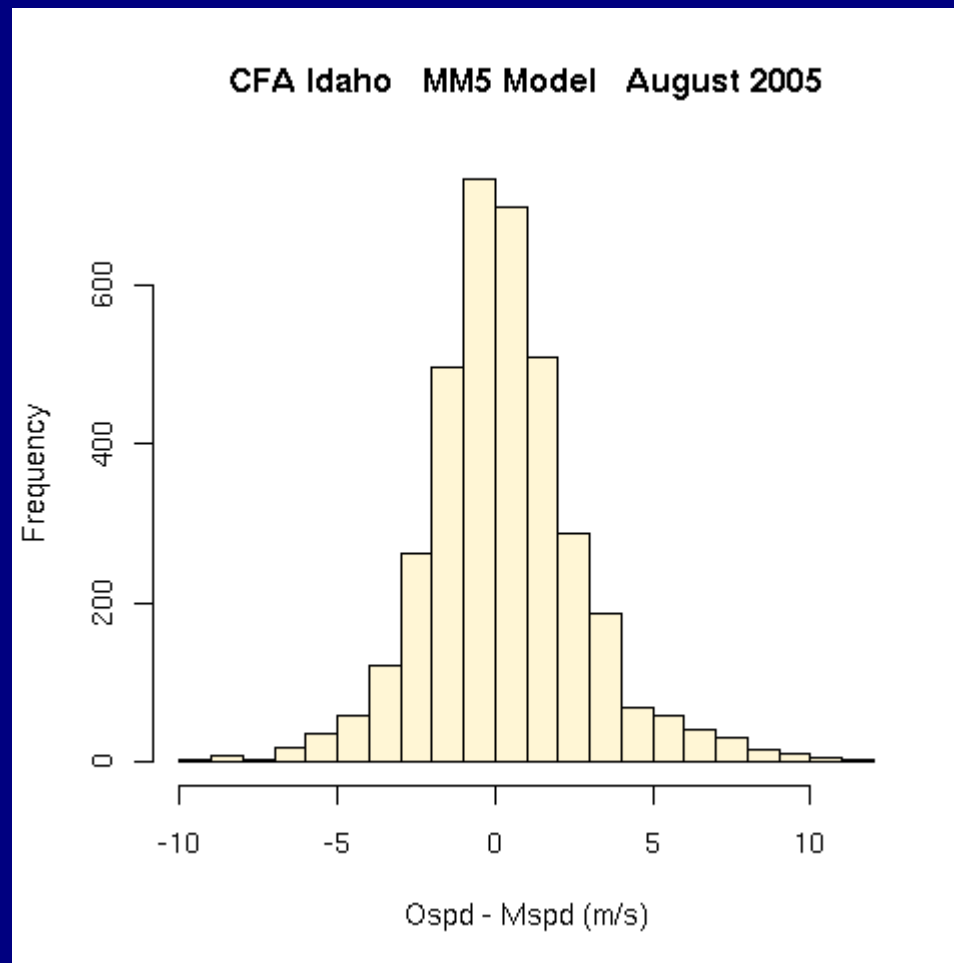
Total uncertainty

Model physics
errors

Model input
uncertainty
(ensemble fcst)

Mean value theorem: $p(F|O_a) = p(F|\hat{M})$

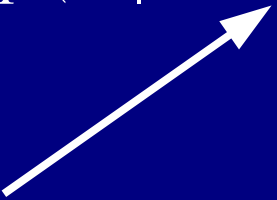
Observed Speeds vs MM5 Forecast



Effect of Recent Observations

O_r = Recent observations not included in model
initialization

$$p(F|O_a, O_r) = \int p(F|M, O_r) p(M|O_a) dM$$



Additional conditional information (O_r)
hopefully decreases uncertainty

Conventional Model Output Statistics (MOS)

→ Normal linear model

$$y_i = \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_k x_{ik} + \epsilon_i$$

or

$$\mathbf{y} = \mathbf{X} \boldsymbol{\beta} + \boldsymbol{\epsilon}$$

- \mathbf{X} variables are combination of model estimates and recent observations
- Coefficients $\boldsymbol{\beta}$ are different for each station
- Direct attempt to model $p(F|M, O_r)$

Example MOS Equation

12 Z wind speed at St. Louis
Glahn and Lowry (1972)

$$S_f^{12} = 1.576 + 0.239 S_m^{12} + 0.175 S_o^{07} - 0.040 V_m^{12} + 0.027 U_m^{12}$$

Combines both 12 Z model output and
recent observations at 7 Z

Urban MOS

$$y = X\beta + \epsilon$$

- y = urban dispersion parameters (skimming flow, turbulence levels)
- X = model outputs (RUC) + local observations (e.g., private network data)
- May use blend of many local observations.
- Utility of private network data demonstrated by improvements in regression

Predictive Distribution

- Predictive distribution for y is multivariate t
- Provides both point estimate and prediction uncertainty
- Standard MOS usage focuses only on point estimate

Advantages

- Conceptually simple
- Linked to private network data
- NOAA operations familiar with MOS approach
- Uses statistics to account for physics not resolved in model
- Predictive distribution gives an estimate of uncertainty in forecast

Disadvantages

- Might need a lot of different regression equations
- Difficult finding reasonable number of explanatory variables
- Might need hierarchical structure to account for complications (e.g., β a function of hyperparameters)
- Purists may criticize lack of physics in regression